# Image and Video Processing (Spring 2023)

# **Assignment 2: Local Operations**

Apr 1, 2023

#### 1 Spatial Filtering [2 points]

Consider the spatial filter H given by:

$$H = \left| \begin{array}{ccc} -1 & -2 & 0 \\ -2 & 0 & 3 \\ 0 & 3 & 1 \end{array} \right|,$$

and an image *I* stored in 6-bit integer format to which the filter *H* is applied.

- 1. Determine the maximum and minimum possible values in  $H \star I$ . Assume no extra normalization and clamping of the values to 6-bit representation during/after filtering.
- 2. Propose a post-filtering pixel transformation that maps back the filtered image to the range of 6-bit integer representation.
- 3. **BONUS**: Is the filter *H* separable? If yes, find its decomposition into smaller filters. If no, find the best separable approximation of this filter. Explain your reasoning. [2 points]

# 2 Combining linear operations [2 points]

During the lecture, we defined an unsharp masking procedure using one filtering step, addition, subtraction, and multiplication. Since all these are linear operations, they can be represented using one linear filter. Given the parameters of the unsharp masking, e.g.,  $\gamma$  and  $\sigma$  of the Gaussian filter, compute one filter kernel that, when applied to an image, gives the unsharp masking effect. For the computation, please assume specific values of the  $\gamma$  and  $\sigma$  and provide the resulting kernel in the report. Additionally, provide the derivation steps and one image with your kernel applied. You can use MATLAB for the calculations but present the steps of the derivation in the report. Remember to choose an appropriate size for the kernel.

### 3 Morphological Operations A [2 points]

The input binary image in Figure 1 represents the greek letter  $\theta$ . What morphological operation(s) would you apply to the input image to obtain an image representing the number zero (see the output image in Figure 1) in the least number of morphological operations? Consider that you can use only **ONE** of the four structuring elements specified in Figure 1 (the 'x' shows the central pixel), but you can use it in as many morphological operations (i.e., erosions/dilations) as needed.

### 4 Morphological Operations B [2 point]

Let B be a structuring element of arbitrary size, such that only one pixel (at any position) in the element is the foreground pixel (value = 1), and the rest are background pixels (value = 0). Let I be a binary image representing background and foreground pixels. What will happen to the foreground pixels, if we apply erosion to image I using the structuring element B? How does this result depend on the position of the foreground pixel in B?

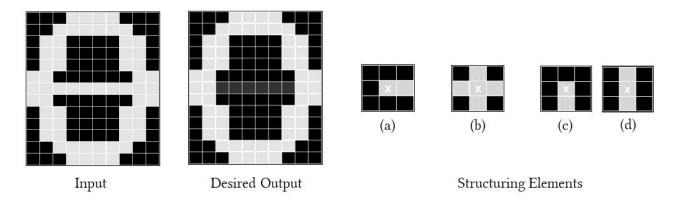


Figure 1: Morphological Operations

#### 5 Linear Motion Blur Filter [4 points]

The Gaussian kernels you have seen in the lecture are radially symmetric, i.e., isotropic. One can make Gaussian filters directional, i.e., anisotropic, by defining standard deviations separately for horizontal and vertical directions. More precisely, given two standard deviations  $\sigma_x$  and  $\sigma_y$  for the horizontal and vertical direction, respectively, the anisotropic Gaussian is defined by:

$$G(x,y) = A \cdot e^{-\left(\frac{x^2}{2\sigma_x^2} + \frac{y^2}{2\sigma_y^2}\right)},\tag{1}$$

where A is the normalizing coefficient that makes the weights sum up to 1.

Additionally, we can rotate the Gaussian function by angle  $\theta$  by applying rotation to the coordinates of the kernel. In practice, you define your rotated Gaussian as:

$$G(x,y) = A \cdot e^{-\left(\frac{x_{\theta}^2}{2\sigma_x^2} + \frac{y_{\theta}^2}{2\sigma_y^2}\right)},\tag{2}$$

where

$$x_{\theta} = x \cdot \cos(\theta) - y \cdot \sin(\theta) \tag{3}$$

$$y_{\theta} = x \cdot \sin(\theta) + y \cdot \cos(\theta). \tag{4}$$

In this exercise, you will simulate a motion blur with the anisotropic Gaussian filter. The idea is to filter an image with a Gaussian filter which is oriented along the motion direction. To this end:

- 1. Implement a function that computes an anisotropic Gaussian kernel for blurring the image along the motion direction. The function should take as an input the angle of motion ( $\theta$ ),  $\sigma_x$ , and  $\sigma_y$ .
- 2. Apply this filter to the provided 'graz.png' image to simulate motion blur at an angle of 45 degrees, as shown in Figure 2(b). Experiment with parameters  $\sigma_x$  and  $\sigma_y$  to obtain the results you like. Include the resulting image and the filter in the report, similarly as demonstrated in Figure 2(b). You can use conv2 function for this exercise.
- 3. **BONUS**: Derive the optimal kernel size for arbitrary values of  $\theta$ ,  $\sigma_x$ , and  $\sigma_y$ . [1 points]

# 6 Iterative filtering [4 points]

Applying a Gaussian filter to an image multiple times results in a similar result to applying one bigger Gaussian filter; see Figure 3. Verify empirically to what extent this is true also for a bilateral filter. For the experiments, you can use the implementation of the bilateral filter provided in MATLAB (imbilatfilt). Note that the two parameters of this function (DegreeOfSmoothing and SpatialSigma) correspond to standard deviation parameters of range and domain kernels discussed during the lecture. Present the results and

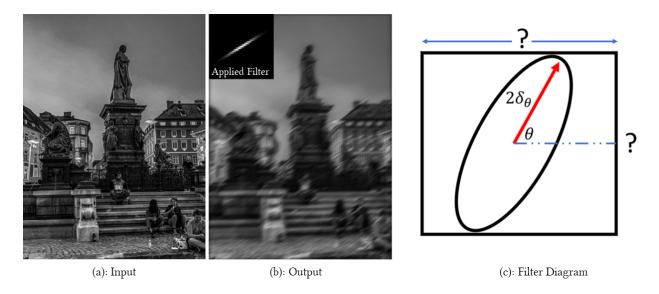


Figure 2: Motion Blurring

observations of your experiments in the report. Please provide an explanation for any differences between the results of applying a bigger bilateral filter once versus applying multiple times a smaller filter.

**BONUS**: Prove that applying two Gaussian filters consecutively to an image is equivalent to applying one Gaussian filter with a bigger  $\sigma$  parameter. Given that the standard deviations of the two Gaussian filters applied sequentially are  $\gamma$  and  $\tau$ , what is the standard deviation of one equivalent Gaussian filter? [1 point]







10x Gaussian filter (standard deviation = 2.0)

Figure 3: Example of applying iterative filtering. The image on the left presents the result of applying one big Gaussian filter. The image on the right presents the result of applying multiple times a much smaller Gaussian filter. Note the similarity of the results.

#### 6.1 Image Stylization [4 points]

Design and implement a filtering procedure that performs image stylization visualized in Figure 4. More precisely, use the knowledge from the lectures on linear and non-linear filtering to develop a sequence of filtering steps that turn the image on the left in Figure 4 into the image on the left. You do not need to obtain the same, but somewhat similar results. However, we want you to focus on two aspects of the stylization you are asked to develop. First, it smoothes the image in a particular way. Second, it adds black lines along image discontinuities. These lines are approximately 2-3 pixels wide. In your report, explain the sequence of filtering steps and provide the intermediate results. For the implementation, you can use any MATLAB in-

build functions which help you implement the filtering procedure. With the assignment, we provide image *delicate\_arch.jpg*, which you should use for this exercise. In-case it is needed, we also provide a sample-script (ColorHandling.m) of how to separate the color from the gray channel; process the gray channel, and then re-colorize the image.





Input image

Output image

Figure 4: Stylization example

#### **Submission**

You should submit one ZIP-file via iCorsi containing:

- All your code in MATLAB appropriately commented, and the processed pictures that you obtained.
- A complete PDF report detailing your solution, partial results for each exercise, and answers to the theoretical questions poised.

Grading will be mostly based on the provided PDF report so we encourage clarity and detailed answers. We recommend using LATEX or Overleaf to write the report. Usage of ChatGPT or any other natural language model is strictly prohibited and will be severely punished.

Solutions must be returned on Apr 21, 2023 via iCorsi3